

Introduction to Machine Learning

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A Few Quotes

- “A breakthrough in machine learning would be worth ten Microsofts”
 - Bill Gates, Chairman, Microsoft

- “Machine learning is the next Internet”
 - Tony Tether, Director, DARPA

- “Machine learning is the hot new thing”
 - John Hennessy, President, Stanford

- “Web rankings today are mostly a matter of machine learning”
 - Prabhakar Raghavan, Dir. Research, Yahoo

- “Machine learning is going to result in a real revolution”
 - Greg Papadopoulos, Former CTO, Sun

- “Machine learning is today’s discontinuity”
 - Jerry Yang, Founder, Yahoo

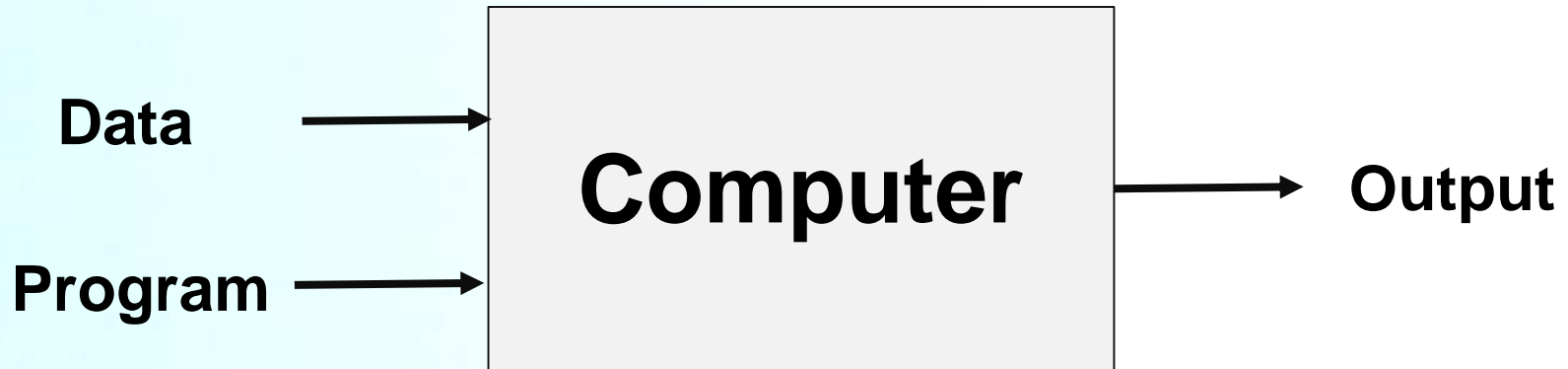
- “Machine learning today is one of the hottest aspects of computer science”
 - Steve Ballmer, CEO, Microsoft

What is Machine Learning?

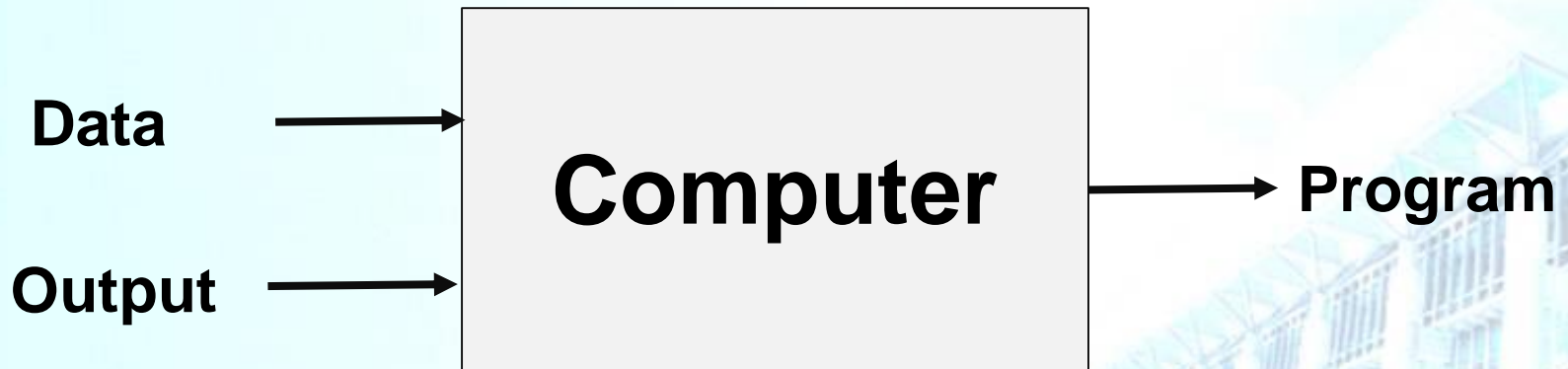
- Automating automation
- Getting computers to program themselves
- Writing software is the bottleneck
- Let the data do the work instead!

Traditional Program vs Machine Learning

■ Traditional program



■ Machine learning



When Do We Use Machine Learning?

- ML is used when
 - Human expertise does not exist
 - Navigating on Mars
 - Humans cannot explain their expertise
 - Walking, Speech recognition
 - Models must be customized
 - Personalized medicine
 - Models are based on huge amounts of data
 - Genomics
 - Solution changes in time
 - Routing on a computer network

More Examples of ML

■ Recognition of patterns

- Facial identities or facial expressions
- Handwritten or spoken words
- Medical images

■ Generation of patterns

- Generating images or motion sequences

■ Recognition of anomalies

- Unusual credit card transactions
- Unusual patterns of sensor readings in a nuclear power plant

■ Prediction

- Future stock prices or currency exchange rates

Application Areas

- Computer vision
- Speech recognition
- Web search
- Computational biology
- Finance
- E-commerce
- Space exploration
- Robotics
- Information extraction
- Social networks
- Debugging
- ...

Machine Learning

- Every machine learning algorithm has three components:
 - Representation
 - Evaluation
 - Optimization

Representation

- Decision trees
- Sets of rules / Logic programs
- Instances
- Graphical models (Bayes/Markov nets)
- Neural networks
- Support vector machines
- Model ensembles
- ...

Evaluation

- Accuracy
- Precision and recall
- Squared error
- Likelihood
- Posterior probability
- Cost / Utility
- Margin
- Entropy
- K-L divergence
- ...

Optimization

- Combinatorial optimization
 - Greedy search

- Convex optimization
 - Gradient descent

- Constrained optimization
 - Linear programming

Types of Learning

- Supervised (inductive) learning
 - Given: training data + desired outputs (labels)

- Unsupervised learning
 - Given: training data (without desired outputs)

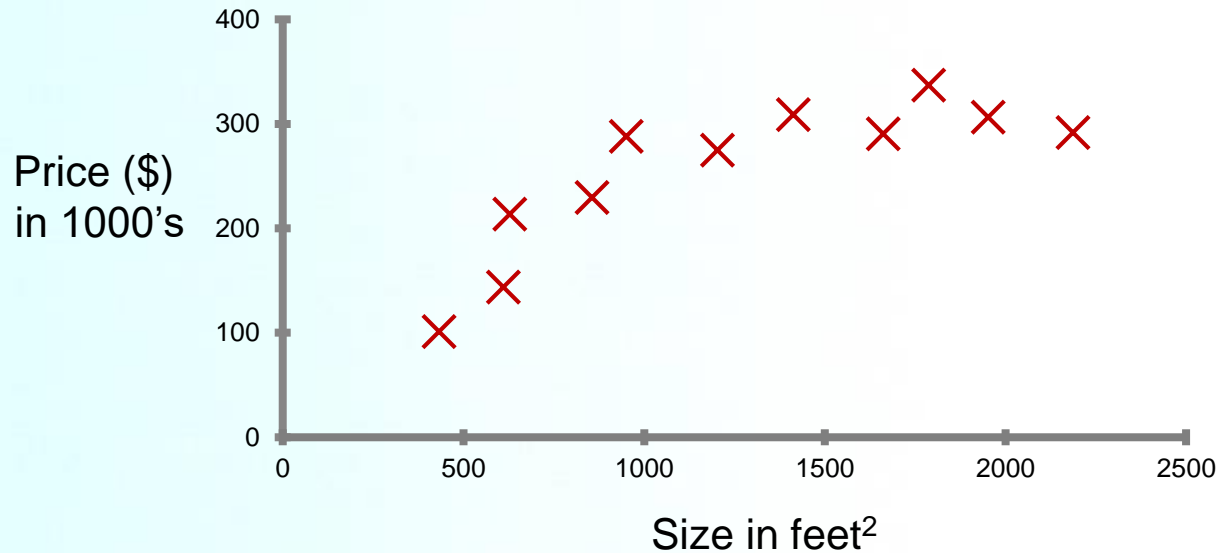
- Semi-supervised learning
 - Given: training data + a few desired outputs

- Reinforcement learning
 - Rewards from sequence of actions

Inductive Learning

- **Given** examples of a function $(x, f(x))$
 - **Predict** function $f(x_{new})$ for a new example x_{new}
 - Discrete $f(x)$: Classification
 - Continuous $f(x)$: Regression
 - $f(x) = \text{Probability}(x)$: Probability estimation

Supervised Learning: Regression



■ Supervised Learning

■ “right answers” given

■ Given $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

- Learn a function $f(x)$ to predict y given x
- y is real-valued → regression

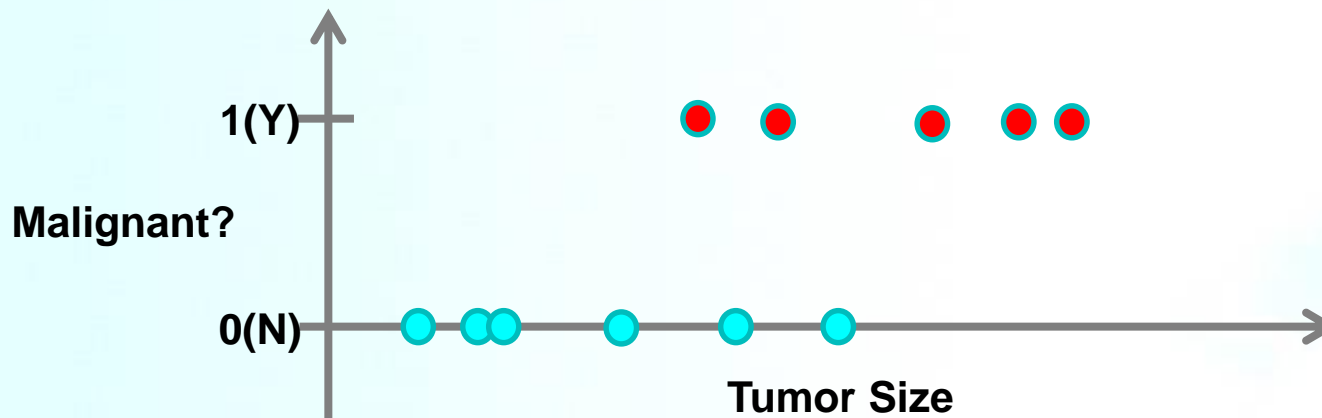
■ Regression

■ Predict continuous valued output (price)

Supervised Learning: Classification

- Given $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$
 - Learn a function $f(x)$ to predict y given x
 - y is categorical \rightarrow classification

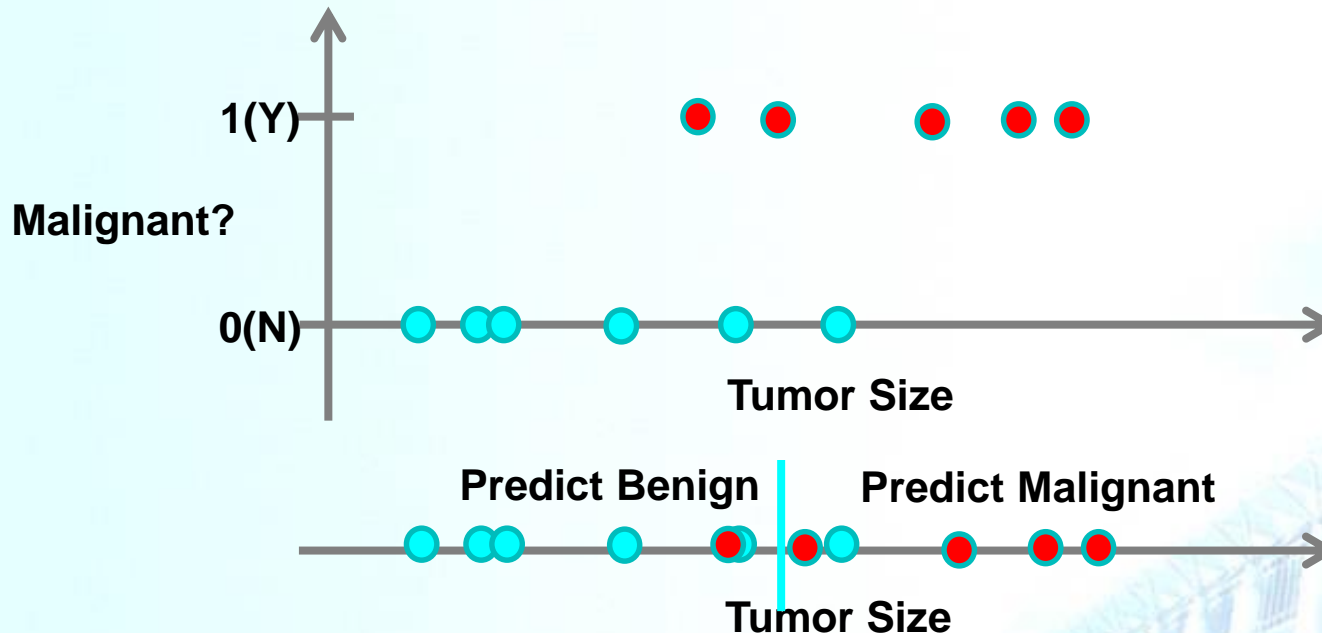
- Classification
 - Discrete valued output (0 or 1)



Supervised Learning: Classification

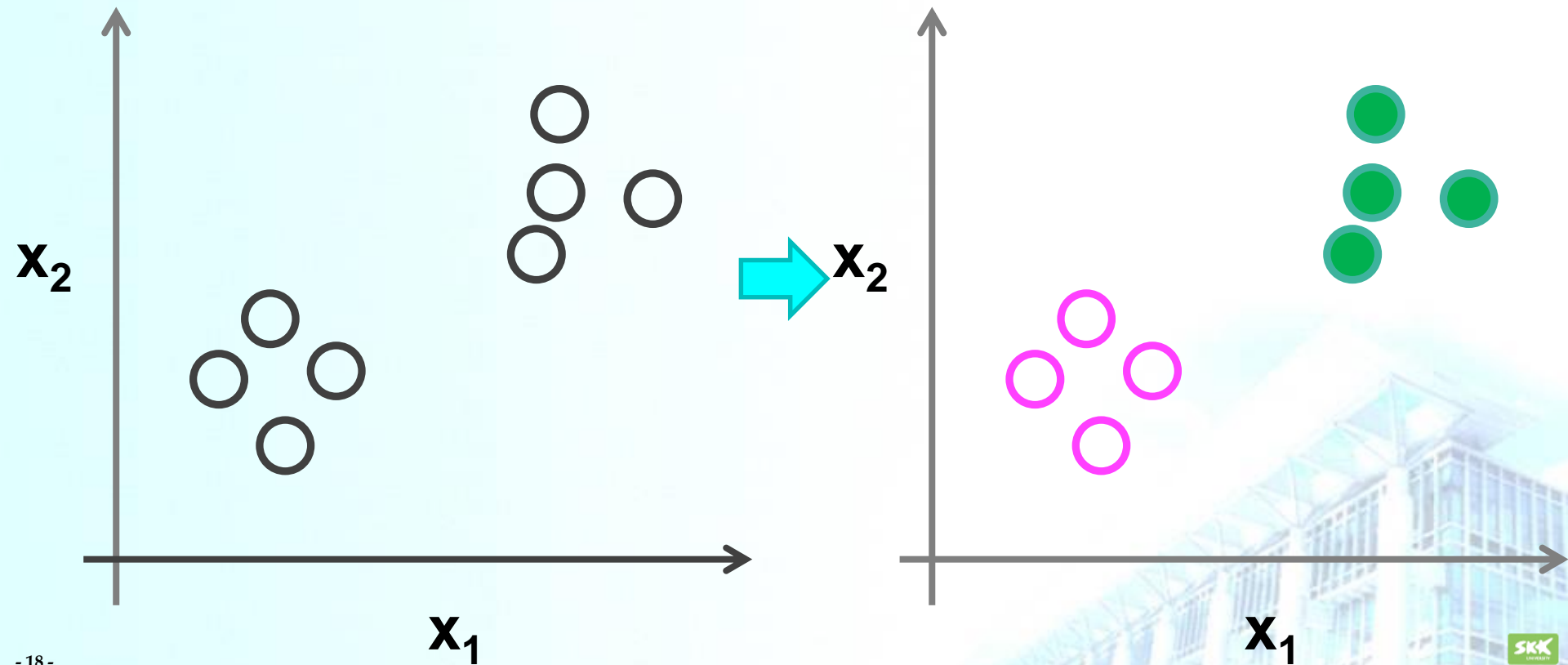
- Given $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$
 - Learn a function $f(x)$ to predict y given x
 - y is categorical \rightarrow classification

- Classification
 - Discrete valued output (0 or 1)



Unsupervised Learning

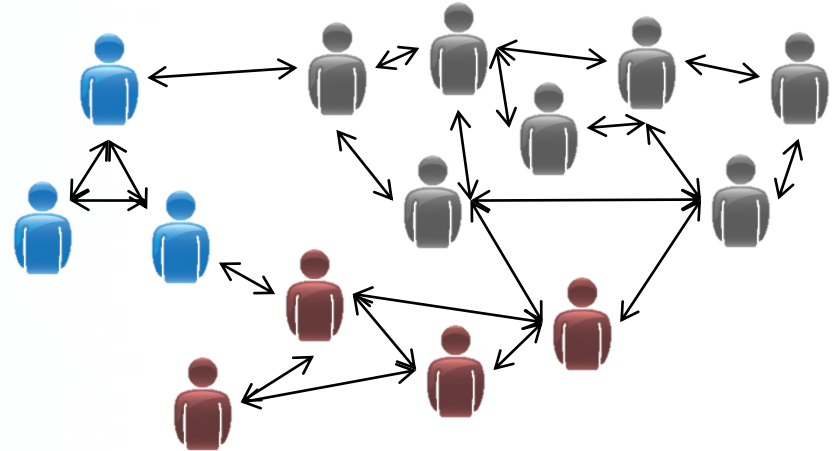
- Given x_1, x_2, \dots, x_n (without labels)
- Output hidden structure behind the x 's
 - clustering



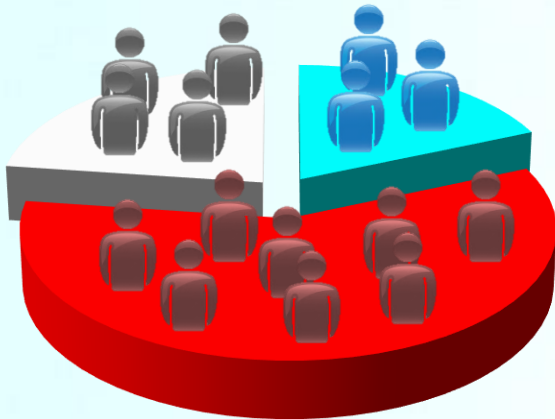
Unsupervised Learning



Organize computing clusters



Social network analysis



Market segmentation

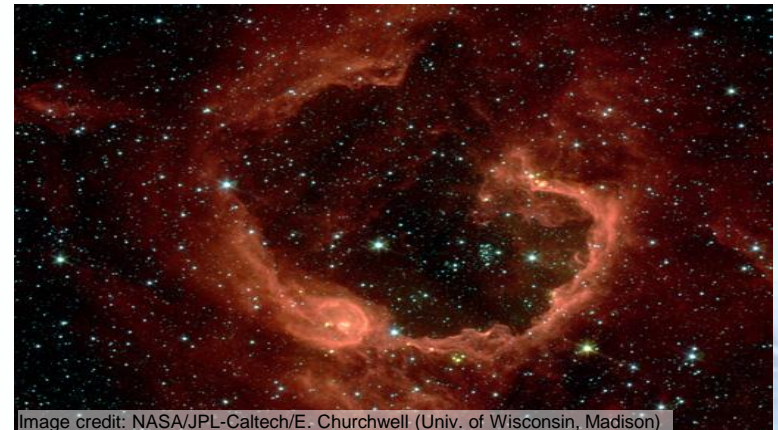


Image credit: NASA/JPL-Caltech/E. Churchwell (Univ. of Wisconsin, Madison)

Astronomical data analysis

Reinforcement Learning

- Given a sequence of states and actions with (delayed) rewards, output a policy
 - Policy is a mapping
from states → actions that tells us what to do in a given state
- Examples:
 - Credit assignment problem
 - Game playing
 - Robot in a maze
 - Balance a pole on your hand

Machine Learning Definition

■ Arthur Samuel (1959)

■ Machine Learning

- Field of study that gives computers the ability to learn without being explicitly programmed.

■ Tom Mitchell (1998)

■ Well-posed Learning Problem:

- A computer program is said to *learn* from experience E with respect to some task T and some performance measure P , if its performance on T , as measured by P , improves with experience E .

■ Machine Learning

- Study of algorithms that
 - improve their performance P
 - at some task T
 - with experience E .

Machine Learning Definition

- Suppose our email program watches which emails we do or do not mark as spam, and based on that learns how to better filter spam.
 - Task T
 - Classifying emails as spam or not spam
 - Experience E
 - Watching that we label emails as spam or not spam.
 - Performance P
 - The number (or fraction) of emails correctly classified as spam/not spam.

Machine Learning Definition

- Improve on task T , wrt performance metric P , based on experience E
 - T : Playing checkers
 - P : Percentage of games won against an arbitrary opponent
 - E : Playing practice games against itself

- T : Recognizing hand-written words
 - P : Percentage of words correctly classified
 - E : Database of human-labeled images of handwritten words

- T : Driving on four-lane highways using vision sensors
 - P : Average distance traveled before a human-judged error
 - E : A sequence of images and steering commands recorded while observing a human driver

References

- Andrew Ng, <https://www.coursera.org/learn/machine-learning>
- Eric Eaton, <https://www.seas.upenn.edu/~cis519>
- Pedro Domingos, <https://courses.cs.washington.edu/courses/cse446/14wi/>